

Supplementary materials for

Heavy Ball Momentum Does Not Always Accelerate SGD

.1. Proof of Lemma 3.1

We need to exploit the eigenvalues of \mathcal{T} , i.e., the complex number λ satisfying

$$\det \begin{pmatrix} (\lambda - 1 - \beta)\mathbf{I} + \gamma\mathbf{A} & \beta\mathbf{I} \\ -\mathbf{I} & \lambda\mathbf{I} \end{pmatrix} = 0.$$

Then we have

$$\begin{aligned} \det \begin{pmatrix} (\lambda + \frac{\beta}{\lambda} - 1 - \beta)\mathbf{I} + \gamma\mathbf{A} & \mathbf{0} \\ -\mathbf{I} & \lambda\mathbf{I} \end{pmatrix} &= 0 \\ \implies \det((\lambda + \frac{\beta}{\lambda})\mathbf{I} - [(1 + \beta)\mathbf{I} - \gamma\mathbf{A}]) &= 0. \end{aligned}$$

If λ^* is a eigenvalue of \mathbf{A} , we just need to consider

$$\lambda + \frac{\beta}{\lambda} = (1 + \beta)\mathbf{I} - \gamma\lambda^*. \quad (8)$$

Let $\mathbf{U} := [\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_d]$ be the eigenvectors of \mathbf{A} , it then holds

$$\langle \mathbf{u}_i, \mathbf{u}_j \rangle = 0 \text{ if } i \neq j,$$

since \mathbf{A} is symmetry positive definite. It is easy to see that $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_d$ are the eigenvectors of $(1 + \beta)\mathbf{I} - \gamma\mathbf{A}$. Let λ_i be the i th eigenvalue of \mathbf{A} , we can see

$$(1 + \beta - \gamma\lambda_i)^2 - 4\beta \leq (1 + \beta - \gamma\nu)^2 - 4\beta \leq 0.$$

Thus, we define $\overline{\lambda}_i$ and $\underline{\lambda}_i$ as follows

$$\begin{aligned} \overline{\lambda}_i &:= \frac{(1 + \beta - \gamma\lambda_i) + \sqrt{4\beta - (1 + \beta - \gamma\lambda_i)^2}\mathbf{i}}{2}, \\ \underline{\lambda}_i &:= \frac{(1 + \beta - \gamma\lambda_i) - \sqrt{4\beta - (1 + \beta - \gamma\lambda_i)^2}\mathbf{i}}{2}, \end{aligned}$$

where $\mathbf{i}^2 = -1$. Direct calculating gives us

$$\mathcal{T} \begin{pmatrix} \overline{\lambda}_i \mathbf{u}_i \\ \mathbf{u}_i \end{pmatrix} = \overline{\lambda}_i \begin{pmatrix} \overline{\lambda}_i \mathbf{u}_i \\ \mathbf{u}_i \end{pmatrix}, \mathcal{T} \begin{pmatrix} \underline{\lambda}_i \mathbf{u}_i \\ \mathbf{u}_i \end{pmatrix} = \underline{\lambda}_i \begin{pmatrix} \underline{\lambda}_i \mathbf{u}_i \\ \mathbf{u}_i \end{pmatrix}.$$

Therefore, all the eigenvectors of \mathcal{T} can be written as

$$\left\{ \begin{pmatrix} \overline{\lambda}_i \mathbf{u}_i \\ \mathbf{u}_i \end{pmatrix}, \begin{pmatrix} \underline{\lambda}_i \mathbf{u}_i \\ \mathbf{u}_i \end{pmatrix} \right\}_{1 \leq i \leq d}.$$

If $i \neq j$, we have

$$\left\langle \begin{pmatrix} \overline{\lambda}_i \mathbf{u}_i \\ \mathbf{u}_i \end{pmatrix}, \begin{pmatrix} \underline{\lambda}_j \mathbf{u}_j \\ \mathbf{u}_j \end{pmatrix} \right\rangle = 0.$$

Note that \mathcal{T} has $2d$ different eigenvectors since $\beta = [1 - \sqrt{\gamma\nu}]^2 + \varrho$, $\overline{\lambda}_i \neq \underline{\lambda}_i$. Denote that

$$\overline{\Lambda} := \text{Diag}(\overline{\lambda}_1, \overline{\lambda}_2, \dots, \overline{\lambda}_d), \underline{\Lambda} := \text{Diag}(\underline{\lambda}_1, \underline{\lambda}_2, \dots, \underline{\lambda}_d).$$

We then construct the following matrix

$$\mathcal{U} := \begin{pmatrix} \overline{\Lambda}\mathbf{U} & \underline{\Lambda}\mathbf{U} \\ \mathbf{U} & \mathbf{U} \end{pmatrix}.$$

It is known that \mathcal{T} can be decomposed as

$$\mathcal{T} = \mathcal{U}^{-1} \begin{bmatrix} \bar{\Lambda} & \\ & \underline{\Lambda} \end{bmatrix} \mathcal{U}. \quad (9)$$

Therefore,

$$\mathcal{T}^k = \mathcal{U}^{-1} \underbrace{\begin{bmatrix} \bar{\Lambda}^k & \\ & \underline{\Lambda}^k \end{bmatrix}}_{:=\Lambda} \mathcal{U}.$$

We are then led to

$$\|\mathcal{T}^k\| = \|\mathcal{U}\Lambda^k\mathcal{U}^{-1}\| \leq \|\mathcal{U}\|_F \|\mathcal{U}^{-1}\|_F \cdot 2d|\lambda_{\max}|^k, \quad (10)$$

where we use the fact that $\|\mathbf{M}\mathbf{N}\|_F \leq \max\{\|\mathbf{M}\|_F\|\mathbf{N}\|, \|\mathbf{M}\|\|\mathbf{N}\|_F\}$. When $\beta = (1 - \sqrt{\gamma\nu})^2 + \varrho$ and $0 < \varrho \ll \epsilon$,

$$|\lambda_{\max}| \leq 1 - \sqrt{\gamma\nu} + \varrho.$$

Direct calculation yields

$$\mathcal{U}^{-1} := \begin{pmatrix} \mathbf{U}^\top (\bar{\Lambda} - \underline{\Lambda})^{-1} & -\mathbf{U}^\top (\bar{\Lambda} - \underline{\Lambda})^{-1} \underline{\Lambda} \\ -\mathbf{U}^\top (\bar{\Lambda} - \underline{\Lambda})^{-1} & \mathbf{U}^\top (\bar{\Lambda} - \underline{\Lambda})^{-1} \bar{\Lambda} \end{pmatrix}.$$

From the form of $\bar{\Lambda}, \underline{\Lambda}$, we have

$$[(\bar{\Lambda} - \underline{\Lambda})^{-1}]_{i,i} = [(\bar{\Lambda} - \underline{\Lambda})_{i,i}]^{-1} = -\frac{1}{\sqrt{4\beta - (1 + \beta - \gamma\lambda_i)^2}} \mathbf{i} \approx -\frac{1}{\sqrt{\gamma\nu}} \mathbf{i}.$$

That means

$$(\bar{\Lambda} - \underline{\Lambda})^{-1} \approx \frac{-\mathbf{i}}{\sqrt{\gamma\nu}} \mathbf{I}.$$

Noticing that β is very closed to 1 and ϵ is very small,

$$\underline{\Lambda} \approx \mathbf{I}, \quad \bar{\Lambda} \approx \mathbf{I}.$$

Turning back to \mathcal{U} and \mathcal{U}^{-1} , we see that $\|\mathcal{U}\|_F = \mathcal{O}(1)$ and $\|\mathcal{U}^{-1}\|_F = \Theta(\frac{1}{\sqrt{\gamma\nu}})$.

.2. Proof of Lemma 3.2

If $\beta = 1 - \Theta(\gamma^\tau)$ and $\tau \geq 1$, we have $\beta \geq (1 - \sqrt{\gamma\nu})^2$ when γ is small, it holds $(1 + \beta) - \gamma\nu \leq 2\sqrt{\beta}$, the equation (8) has complex roots whose norms are both β . Thus

$$|\lambda_i| = \sqrt{\beta} \geq 1 - \Theta(\gamma^\tau), \quad 1 \leq i \leq 2d.$$

With such a choice, we still have $(\bar{\Lambda} \approx \underline{\Lambda})^{-1}$, and $\|\mathcal{U}\|_F = \Theta(1)$. Here, the norm $\|\cdot\|_F$ and $\|\cdot\|$ are taken on the complex domain. Let $\xi = (\xi_1 \in \mathbb{R}^d, \mathbf{0})$ and $\xi_1 \sim \mathcal{E}$. Denote $\bar{\xi} := \mathcal{U}\xi = \begin{bmatrix} \bar{\Lambda}\mathbf{U}\xi_1 \\ \underline{\Lambda}\mathbf{U}\xi_1 \end{bmatrix}$. We then have

$$\begin{aligned} \mathbb{E}\|\mathcal{T}^k\xi\|^2 &= \mathbb{E}\|\mathcal{U}^{-1}\Lambda\bar{\xi}\|^2 \geq \mathbb{E}\|\Lambda\bar{\xi}\|^2 / \|\mathcal{U}\|_F^2 = \mathbb{E}\left\| \begin{bmatrix} \bar{\Lambda}^{k+1}\mathbf{U}\xi_1 \\ \underline{\Lambda}^k\mathbf{U}\xi_1 \end{bmatrix} \right\|^2 / \|\mathcal{U}\|_F^2 \\ &\geq \mathbb{E}\|\underline{\Lambda}^k\mathbf{U}\xi_1\|^2 / \|\mathcal{U}\|_F^2 \geq [1 - \Theta(\gamma^\tau)]^{2k} \mathbb{E}\|\mathbf{U}\xi_1\|^2 / \|\mathcal{U}\|_F^2, \end{aligned}$$

Since $\text{Tr}(\mathbf{U}\Sigma\mathbf{U}^\top) = \text{Tr}(\mathbf{U}^\top\mathbf{U}\Sigma) = \text{Tr}(\Sigma)$, we then get

$$\mathbb{E}\|\mathbf{U}\xi_1\|^2 = \text{Tr}(\Sigma).$$

Therefore, we have

$$\mathbb{E}\|\mathcal{T}^k\xi\|^2 \geq \text{Tr}(\Sigma) / \|\mathcal{U}\|_F^2 [1 - \Theta(\gamma^\tau)]^{2k} = \Theta(1) [1 - \Theta(\gamma^\tau)]^{2k}.$$

3. Proof of Lemma 3.5

Let λ_i be the i th eigenvalue of \mathbf{A} and $0 \leq \beta \leq \beta_0 < 1, 1 - \beta_0 \gg \epsilon$, we can see

$$(1 + \beta - \gamma\lambda_i)^2 - 4\beta \geq (1 + \beta - \gamma L)^2 - 4\beta \geq 0.$$

Thus, we define $\overline{\lambda_i}$ and $\underline{\lambda_i}$ as follows

$$\begin{aligned}\overline{\lambda_i} &:= \frac{(1 + \beta - \gamma\lambda_i) + \sqrt{(1 + \beta - \gamma\lambda_i)^2 - 4\beta}}{2}, \\ \underline{\lambda_i} &:= \frac{(1 + \beta - \gamma\lambda_i) - \sqrt{(1 + \beta - \gamma\lambda_i)^2 - 4\beta}}{2}.\end{aligned}$$

Then, we have

$$[(\overline{\Lambda} - \underline{\Lambda})^{-1}]_{i,i} = ([\overline{\Lambda} - \underline{\Lambda}]_{i,i})^{-1} = -\frac{1}{\sqrt{(1 + \beta - \gamma\lambda_i)^2 - 4\beta}} \approx \frac{1}{1 - \beta}.$$

That means

$$\|(\overline{\Lambda} - \underline{\Lambda})^{-1}\| = \Theta\left(\frac{1}{1 - \beta_0}\right), \quad \|\overline{\Lambda}\| = \mathcal{O}(1), \quad \|\underline{\Lambda}\| = \mathcal{O}(1).$$

On the other hand,

$$\frac{(1 + \beta - \gamma\lambda_i) + \sqrt{(1 + \beta - \gamma\lambda_i)^2 - 4\beta}}{2} \leq 1 - \frac{\gamma\lambda_i}{1 - \beta} + C_3\epsilon^2,$$

which means

$$|\lambda_{\max}| \leq 1 - \frac{\gamma\nu}{1 - \beta} + C_3\epsilon^2.$$

4. Proof of Theorem 3.3

Let

$$\mathbf{y}^k := \begin{bmatrix} \mathbf{w}^k - \mathbf{w}^* \\ \mathbf{w}^{k-1} - \mathbf{w}^* \end{bmatrix} \in \mathbb{R}^{2d}.$$

According to the fact that $\nabla R_S(\mathbf{w}^*) = \mathbf{0}$, we have

$$\begin{aligned}\mathbf{w}^{k+1} - \mathbf{w}^* &= \mathbf{w}^k - \mathbf{w}^* - \gamma(\nabla R_S(\mathbf{w}^k) - \nabla R_S(\mathbf{w}^*)) + \beta(\mathbf{w}^k - \mathbf{w}^*) - \beta(\mathbf{w}^{k-1} - \mathbf{w}^*) + \gamma(\mathbf{g}^k - \nabla R_S(\mathbf{w}^k)) \\ &= \mathbf{w}^k - \mathbf{w}^* - \gamma\mathbf{A}(\mathbf{w}^k - \mathbf{w}^*) + \beta(\mathbf{w}^k - \mathbf{w}^*) - \beta(\mathbf{w}^{k-1} - \mathbf{w}^*) + \gamma(\mathbf{g}^k - \nabla R_S(\mathbf{w}^k)),\end{aligned}$$

where $\mathbf{A} := \nabla^2 R_S$. Then SHB can be reformulated as

$$\mathbf{y}^{k+1} = \mathcal{T}\mathbf{y}^k + \mathbf{e}^k,$$

where $\mathbf{e}^k := \begin{pmatrix} \gamma(\mathbf{g}^k - \nabla R_S(\mathbf{w}^k)) \\ \mathbf{0} \end{pmatrix}$. It is easy to see that \mathbf{A} is symmetry positive definite due to the quadratic property of R_S . We then have

$$\mathbf{y}^{k+1} = \mathcal{T}^k \mathbf{y}^1 + \sum_{i=1}^k \mathcal{T}^{k-i} \mathbf{e}^i.$$

Using the fact that $\mathbb{E}\langle \mathbf{e}^i, \mathbf{e}^j \rangle = 0$ if $i \neq j$, we have

$$\mathbb{E}\|\mathbf{y}^{k+1}\|^2 = \mathbb{E}\|\mathcal{T}^k \mathbf{y}^1 + \sum_{i=1}^k \mathcal{T}^{k-i} \mathbf{e}^i\|^2 = \mathbb{E}\|\mathcal{T}^k \mathbf{y}^1\|^2 + \sum_{i=1}^k \|\mathcal{T}^{k-i} \mathbf{e}^i\|^2. \quad (11)$$

With Lemma 3.1, it follows

$$\mathbb{E}\|\mathbf{w}^K - \mathbf{w}^*\|^2 \leq \mathbb{E}\|\mathbf{y}^K\|^2 \leq \frac{C_1^2}{\gamma\nu} [1 - \sqrt{\gamma\nu}]^{2K} \|\mathbf{y}^1\|^2 + \frac{\gamma C_1^2 \sigma^2}{\nu} \sum_{i=1}^K [1 - \sqrt{\gamma\nu}]^{2K-2i}.$$

When γ is small, $\sum_{i=1}^K [1 - \sqrt{\gamma\nu}]^{2K-2i} \leq \frac{1}{\sqrt{\gamma\nu}}$, we then proved the result.

.5. Proof of Theorem 3.4

Noticing that with Lemma 3.2, it holds $\mathbb{E}\|\mathcal{T}^{K-i}\mathbf{e}^i\|^2 \geq C_2 \cdot \gamma^2(1 - \Theta(\gamma^\tau))^{2K-2i}$. Stating from (11), we are then led to

$$\mathbb{E}\|\mathbf{y}^K\|^2 \geq \mathbb{E}\|\mathcal{T}^k\mathbf{y}^1\|^2 + C_2\gamma^2 \sum_{i=1}^K [1 - \Theta(\gamma^\tau)]^{2K-2i} = \Theta(\gamma^{2-\tau}).$$

The above equation indicates that

$$\mathbb{E}\|\mathbf{w}^K - \mathbf{w}^*\|^2 + \mathbb{E}\|\mathbf{w}^{K-1} - \mathbf{w}^*\|^2 \geq \Theta(\gamma^{2-\tau}). \quad (12)$$

According to (12), if we set $\gamma = \Theta(\epsilon)$, the lower bound is in the order of $\Theta(\epsilon^{2-\tau})$.

.6. Proof of Theorem 3.6

Note that (11) still holds. With Lemma 6, we have

$$\mathbb{E}\|\mathbf{w}^K - \mathbf{w}^*\|^2 \leq C_4^2 [1 - \frac{\gamma\nu}{1-\beta} + C_3\epsilon^2]^{2K} \|\mathbf{y}^1\|^2 + \gamma^2 C_4^2 \sigma^2 \sum_{i=1}^K [1 - \frac{\gamma\nu}{1-\beta} + C_3\epsilon^2]^{2K-2i}.$$

When $\Theta(\epsilon)$ and ϵ is small,

$$\gamma^2 C_4^2 \sigma^2 \sum_{i=1}^K [1 - \frac{\gamma\nu}{1-\beta} + C_3\epsilon^2]^{2K-2i} = C_4^2 \sigma^2 \frac{1-\beta}{\nu} \gamma + \mathcal{O}(\epsilon^2) = \mathcal{O}(\epsilon).$$

If $\mathbb{E}\|\mathbf{w}^K - \mathbf{w}^*\|^2 \leq \epsilon$, we then have

$$[1 - \frac{\gamma\nu}{1-\beta} + C_3\epsilon^2]^{2K} = \mathcal{O}(\epsilon).$$

The worst case is then $\mathcal{O}(\frac{\ln \frac{1}{\epsilon}}{\frac{\gamma\nu}{1-\beta} - C_3\epsilon^2}) = \tilde{\mathcal{O}}(\frac{1-\beta}{\epsilon\nu})$.